```
import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
[ ]: ak = sns.load_dataset('car_crashes')
     ak
[]:
         total
                 speeding
                           alcohol
                                     not_distracted no_previous
                                                                     ins_premium \
          18.8
                    7.332
                              5.640
                                              18.048
                                                            15.040
                                                                          784.55
          18.1
     1
                    7.421
                              4.525
                                              16.290
                                                            17.014
                                                                         1053.48
     2
          18.6
                    6.510
                              5.208
                                              15.624
                                                            17.856
                                                                          899.47
     3
          22.4
                    4.032
                              5.824
                                              21.056
                                                            21.280
                                                                          827.34
     4
          12.0
                    4.200
                              3.360
                                              10.920
                                                            10.680
                                                                          878.41
     5
          13.6
                    5.032
                              3.808
                                              10.744
                                                            12.920
                                                                          835.50
     6
          10.8
                    4.968
                              3.888
                                               9.396
                                                             8.856
                                                                         1068.73
     7
          16.2
                    6.156
                              4.860
                                              14.094
                                                            16.038
                                                                         1137.87
     8
           5.9
                    2.006
                              1.593
                                               5.900
                                                             5.900
                                                                         1273.89
     9
          17.9
                    3.759
                              5.191
                                              16.468
                                                            16.826
                                                                         1160.13
     10
          15.6
                    2.964
                              3.900
                                                                          913.15
                                              14.820
                                                            14.508
     11
          17.5
                    9.450
                              7.175
                                              14.350
                                                            15.225
                                                                          861.18
     12
          15.3
                    5.508
                              4.437
                                              13.005
                                                            14.994
                                                                          641.96
     13
          12.8
                    4.608
                              4.352
                                              12.032
                                                            12.288
                                                                          803.11
     14
          14.5
                    3.625
                              4.205
                                              13.775
                                                            13.775
                                                                          710.46
     15
          15.7
                    2.669
                              3.925
                                              15.229
                                                            13.659
                                                                          649.06
     16
          17.8
                    4.806
                              4.272
                                              13.706
                                                            15.130
                                                                          780.45
     17
          21.4
                    4.066
                              4.922
                                              16.692
                                                            16.264
                                                                          872.51
     18
          20.5
                              6.765
                    7.175
                                              14.965
                                                            20.090
                                                                         1281.55
     19
                              4.530
          15.1
                    5.738
                                              13.137
                                                            12.684
                                                                          661.88
     20
          12.5
                    4.250
                              4.000
                                               8.875
                                                            12.375
                                                                         1048.78
     21
           8.2
                              2.870
                    1.886
                                               7.134
                                                             6.560
                                                                         1011.14
     22
          14.1
                    3.384
                              3.948
                                              13.395
                                                            10.857
                                                                         1110.61
     23
           9.6
                    2.208
                              2.784
                                               8.448
                                                             8.448
                                                                          777.18
     24
          17.6
                    2.640
                              5.456
                                               1.760
                                                            17.600
                                                                          896.07
```

25	16.1	6.923	5.474	14.812	13.524	790.32
26	21.4	8.346	9.416	17.976	18.190	816.21
27	14.9	1.937	5.215	13.857	13.410	732.28
28	14.7	5.439	4.704	13.965	14.553	1029.87
29	11.6	4.060	3.480	10.092	9.628	746.54
30	11.2	1.792	3.136	9.632	8.736	1301.52
31	18.4	3.496	4.968	12.328	18.032	869.85
32	12.3	3.936	3.567	10.824	9.840	1234.31
33	16.8	6.552	5.208	15.792	13.608	708.24
34	23.9	5.497	10.038	23.661	20.554	688.75
35	14.1	3.948	4.794	13.959	11.562	697.73
36	19.9	6.368	5.771	18.308	18.706	881.51
37	12.8	4.224	3.328	8.576	11.520	804.71
38	18.2	9.100	5.642	17.472	16.016	905.99
39	11.1	3.774	4.218	10.212	8.769	1148.99
40	23.9	9.082	9.799	22.944	19.359	858.97
41	19.4	6.014	6.402	19.012	16.684	669.31
42	19.5	4.095	5.655	15.990	15.795	767.91
43	19.4	7.760	7.372	17.654	16.878	1004.75
44	11.3	4.859	1.808	9.944	10.848	809.38
45	13.6	4.080	4.080	13.056	12.920	716.20
46	12.7	2.413	3.429	11.049	11.176	768.95
47	10.6	4.452	3.498	8.692	9.116	890.03
48	23.8	8.092	6.664	23.086	20.706	992.61
49	13.8	4.968	4.554	5.382	11.592	670.31
50	17.4	7.308	5.568	14.094	15.660	791.14

	ins_losses	abbrev
0	145.08	AL
1	133.93	AK
2	110.35	AZ
3	142.39	AR
4	165.63	CA
5	139.91	CO
6	167.02	CT
7	151.48	DE
8	136.05	DC
9	144.18	FL
10	142.80	GA
11	120.92	HI
12	82.75	ID
13	139.15	IL
14	108.92	IN
15	114.47	IA
16	133.80	KS
17	137.13	KY
18	194.78	LA

```
19
         96.57
                    ME
20
        192.70
                    MD
21
        135.63
                    MA
22
        152.26
                    ΜI
23
        133.35
                    MN
24
        155.77
                    MS
25
        144.45
                    МО
26
         85.15
                    МТ
27
        114.82
                    NE
28
        138.71
                    NV
29
        120.21
                    NH
30
        159.85
                    NJ
31
                    NM
        120.75
32
        150.01
                    NY
33
        127.82
                    NC
34
        109.72
                    ND
35
        133.52
                    OH
36
        178.86
                    OK
37
                    OR
        104.61
38
                    PA
        153.86
39
        148.58
                    RΙ
40
        116.29
                    SC
41
         96.87
                    SD
42
                    TN
        155.57
43
        156.83
                    TX
44
        109.48
                    UT
45
        109.61
                    VT
46
        153.72
                    VA
47
        111.62
                    WA
                    WV
48
        152.56
49
        106.62
                    WI
50
        122.04
                    WY
```

[]: ak.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 51 entries, 0 to 50
Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype
0	total	51 non-null	float64
1	speeding	51 non-null	float64
2	alcohol	51 non-null	float64
3	${\tt not_distracted}$	51 non-null	float64
4	no_previous	51 non-null	float64
5	ins_premium	51 non-null	float64
6	ins losses	51 non-null	float64

memory usage: 3.3+ KB []: ak.head() []: total speeding alcohol not_distracted no_previous ins_premium \ 18.8 7.332 5.640 784.55 18.048 15.040 1 18.1 7.421 4.525 16.290 17.014 1053.48 2 18.6 6.510 5.208 15.624 17.856 899.47 22.4 3 4.032 5.824 21.056 21.280 827.34 4 12.0 4.200 3.360 10.920 10.680 878.41 ins_losses abbrev 0 145.08 AL 133.93 AK 1 2 110.35 AZ3 142.39 AR 4 165.63 CA []: ak.head(2) []: total speeding alcohol not_distracted no_previous ins_premium \ 0 18.8 7.332 5.640 18.048 15.040 784.55 18.1 7.421 4.525 16.290 1053.48 1 17.014 ins_losses abbrev 0 145.08 AL 133.93 1 AK []: ak.tail(8) []: total speeding alcohol not_distracted no_previous ins_premium \ 43 19.4 7.760 7.372 17.654 16.878 1004.75 44 11.3 4.859 1.808 9.944 10.848 809.38 45 13.6 4.080 4.080 13.056 12.920 716.20 46 12.7 2.413 3.429 11.049 11.176 768.95 47 10.6 4.452 3.498 8.692 9.116 890.03 48 23.8 8.092 6.664 23.086 20.706 992.61 49 13.8 4.968 4.554 5.382 11.592 670.31 50 17.4 7.308 5.568 14.094 15.660 791.14 ins_losses abbrev 43 156.83 TX 44 109.48 UT 45 109.61 VT 46 153.72 VA

object

abbrev

dtypes: float64(7), object(1)

51 non-null

```
48
              152.56
                         WV
     49
              106.62
                         WI
     50
              122.04
                         WY
[]:
     ak.tail()
[]:
                            alcohol
                                     not_distracted
                                                                     ins_premium \
         total
                 speeding
                                                      no_previous
     46
          12.7
                    2.413
                              3.429
                                              11.049
                                                                          768.95
                                                            11.176
     47
          10.6
                    4.452
                              3.498
                                               8.692
                                                             9.116
                                                                          890.03
     48
                                              23.086
          23.8
                    8.092
                              6.664
                                                            20.706
                                                                          992.61
     49
          13.8
                    4.968
                              4.554
                                               5.382
                                                            11.592
                                                                          670.31
          17.4
                    7.308
     50
                              5.568
                                              14.094
                                                            15.660
                                                                          791.14
         ins losses abbrev
     46
              153.72
                         VA
     47
              111.62
                         WA
                         WV
     48
              152.56
     49
              106.62
                         WI
     50
              122.04
                         WY
[]:
     ak.shape
[]: (51, 8)
[]:
     ak.describe()
[]:
                 total
                         speeding
                                      alcohol
                                                not_distracted
                                                                 no_previous
            51.000000
                        51.000000
                                    51.000000
                                                      51.000000
                                                                    51.000000
     count
     mean
             15.790196
                         4.998196
                                     4.886784
                                                      13.573176
                                                                    14.004882
     std
             4.122002
                         2.017747
                                     1.729133
                                                       4.508977
                                                                     3.764672
     min
             5.900000
                         1.792000
                                     1.593000
                                                       1.760000
                                                                     5.900000
     25%
             12.750000
                         3.766500
                                     3.894000
                                                      10.478000
                                                                    11.348000
     50%
             15.600000
                         4.608000
                                     4.554000
                                                      13.857000
                                                                    13.775000
     75%
             18.500000
                         6.439000
                                     5.604000
                                                                    16.755000
                                                      16.140000
             23.900000
                         9.450000
     max
                                    10.038000
                                                      23.661000
                                                                    21.280000
             ins_premium
                          ins_losses
     count
               51.000000
                            51.000000
                           134.493137
     mean
             886.957647
     std
              178.296285
                            24.835922
     min
             641.960000
                            82.750000
     25%
             768.430000
                           114.645000
     50%
             858.970000
                           136.050000
     75%
             1007.945000
                           151.870000
     max
             1301.520000
                           194.780000
```

47

111.62

WA

```
[]: corr = ak.corr()
corr
```

<ipython-input-12-cd014e0cc39d>:1: FutureWarning: The default value of
numeric_only in DataFrame.corr is deprecated. In a future version, it will
default to False. Select only valid columns or specify the value of numeric_only
to silence this warning.

corr = ak.corr()

```
[]:
                       total speeding
                                         alcohol not_distracted no_previous \
    total
                    1.000000 0.611548 0.852613
                                                        0.827560
                                                                     0.956179
                    0.611548 1.000000 0.669719
                                                                     0.571976
    speeding
                                                        0.588010
    alcohol
                    0.852613 0.669719
                                        1.000000
                                                        0.732816
                                                                     0.783520
    not_distracted 0.827560 0.588010
                                        0.732816
                                                        1.000000
                                                                     0.747307
    no_previous
                    0.956179 0.571976 0.783520
                                                        0.747307
                                                                     1.000000
                   -0.199702 -0.077675 -0.170612
    ins_premium
                                                       -0.174856
                                                                    -0.156895
    ins_losses
                   -0.036011 -0.065928 -0.112547
                                                       -0.075970
                                                                    -0.006359
                                 ins_losses
                     ins_premium
```

```
total
                  -0.199702
                              -0.036011
speeding
                  -0.077675
                              -0.065928
alcohol
                  -0.170612
                              -0.112547
not distracted
                  -0.174856
                              -0.075970
no_previous
                  -0.156895
                              -0.006359
ins premium
                   1.000000
                               0.623116
ins_losses
                   0.623116
                               1.000000
```

```
[]: plt.subplots(figsize = (20,10))
sns.heatmap(corr,annot=True)
```

[]: <Axes: >



[]: ak["total"].value_counts()

[]: 14.1 2 12.8 2 13.6 2 21.4 2 19.4 2 23.9 2 14.9 1 14.7 1 11.6 1 11.2 1 18.4 1 12.3 1 16.8 1 19.9 1 17.6 1 18.2 1 11.1 1 19.5 1 11.3 1 12.7 1 10.6 1 23.8 1 13.8 1

16.1

1

```
9.6
     18.1
              1
     18.6
              1
     22.4
              1
     12.0
              1
     10.8
              1
     16.2
              1
     5.9
              1
     17.9
              1
     15.6
              1
     17.5
              1
     15.3
              1
     14.5
              1
     15.7
              1
     17.8
              1
     20.5
              1
     15.1
              1
     12.5
              1
     8.2
              1
     17.4
              1
     Name: total, dtype: int64
[]: ak.alcohol.value_counts()
[]: 5.208
                2
     5.640
                1
     4.218
                1
     4.704
                1
     3.480
                1
     3.136
                1
     4.968
                1
     3.567
                1
     10.038
                1
     4.794
                1
     5.771
                1
     3.328
                1
     5.642
                1
     9.799
                1
     9.416
                1
     6.402
                1
     5.655
                1
     7.372
                1
     1.808
                1
     4.080
                1
     3.429
                1
     3.498
                1
```

18.8

1

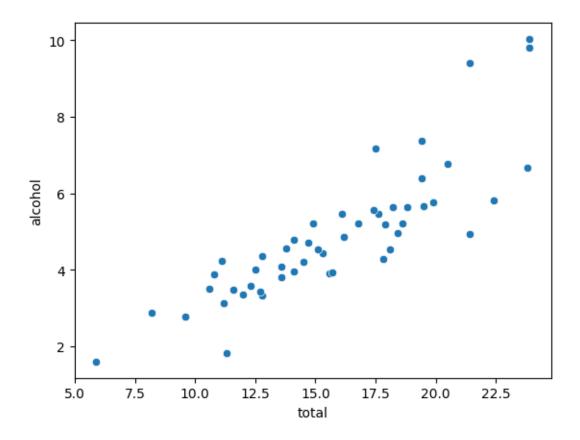
```
6.664
                1
     4.554
                1
     5.215
                1
     5.474
                1
     4.525
                1
     5.456
                1
     5.824
                1
     3.360
                1
     3.808
                1
     3.888
                1
     4.860
                1
     1.593
                1
     5.191
                1
     3.900
                1
     7.175
                1
     4.437
                1
     4.352
                1
     4.205
                1
     3.925
                1
     4.272
                1
     4.922
                1
     6.765
                1
     4.530
                1
     4.000
                1
     2.870
                1
     3.948
                1
     2.784
                1
     5.568
                1
     Name: alcohol, dtype: int64
[]: ak.isnull().any()
[]: total
                        False
     speeding
                        False
     alcohol
                        False
     not_distracted
                        False
     no_previous
                        False
     ins_premium
                        False
     ins_losses
                        False
     abbrev
                        False
     dtype: bool
[]: ak.isnull().sum()
[]: total
                        0
     speeding
                        0
     alcohol
                        0
```

```
not_distracted 0
no_previous 0
ins_premium 0
ins_losses 0
abbrev 0
dtype: int64
```

DATA VISUALIZAION

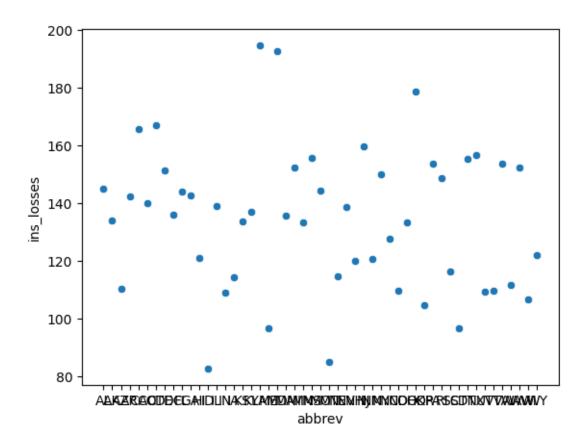
```
[]: sns.scatterplot(x="total",y="alcohol", data=ak)
```

[]: <Axes: xlabel='total', ylabel='alcohol'>



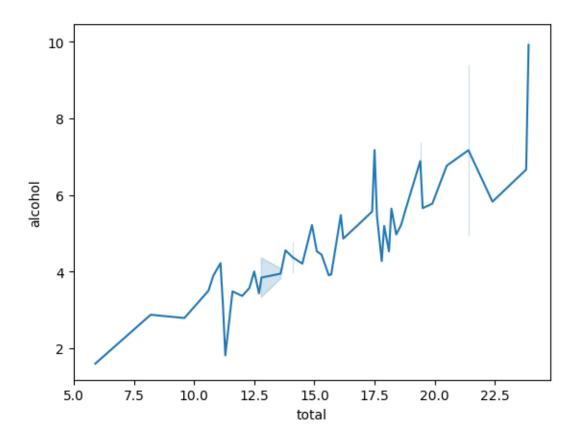
```
[]: sns.scatterplot(x="abbrev",y="ins_losses",data=ak)
```

[]: <Axes: xlabel='abbrev', ylabel='ins_losses'>



```
[]: #Lineplot sns.lineplot(y="alcohol",x="total",data=ak)
```

[]: <Axes: xlabel='total', ylabel='alcohol'>



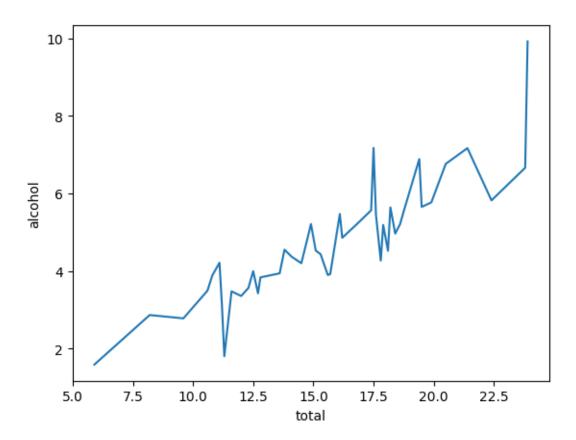
```
[]: #Lineplot sns.lineplot(y="alcohol",x="total",data=ak,ci=None)
```

<ipython-input-21-d491e3384d9c>:2: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

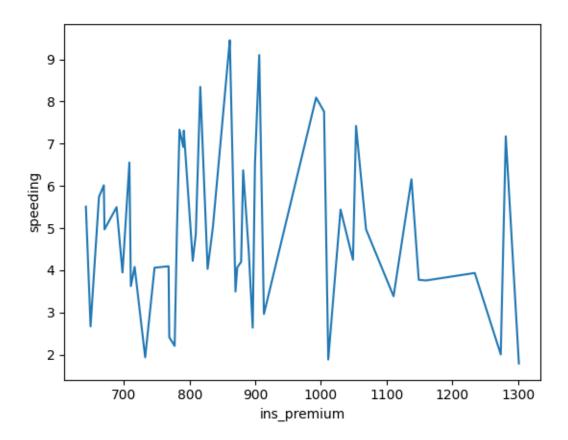
sns.lineplot(y="alcohol",x="total",data=ak,ci=None)

[]: <Axes: xlabel='total', ylabel='alcohol'>



```
[]: sns.lineplot(x="ins_premium",y="speeding",data=ak)
```

[]: <Axes: xlabel='ins_premium', ylabel='speeding'>



[]: #Displot sns.distplot(ak["total"])

<ipython-input-23-18a78abad740>:2: UserWarning:

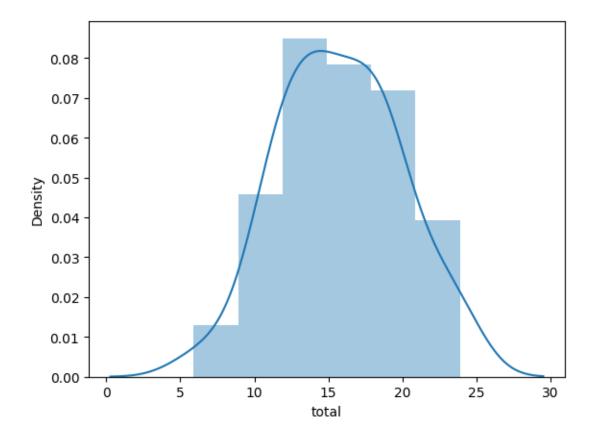
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(ak["total"])

[]: <Axes: xlabel='total', ylabel='Density'>



[]: sns.distplot(ak["not_distracted"])

<ipython-input-24-bb0c8d7ed882>:1: UserWarning:

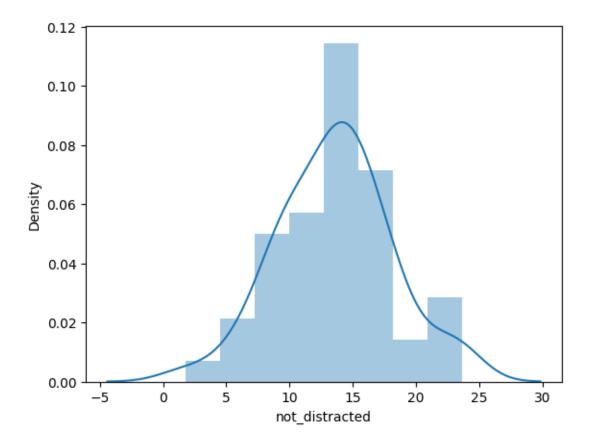
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

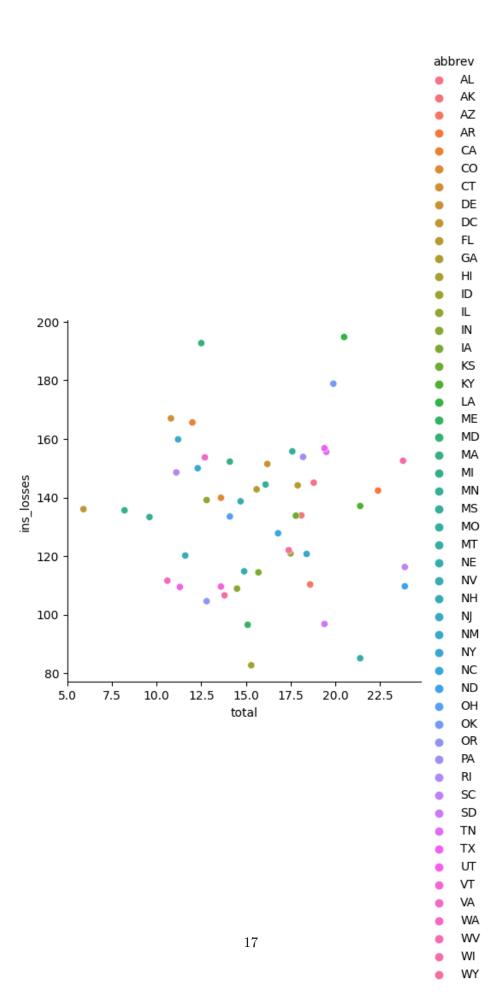
sns.distplot(ak["not_distracted"])

[]: <Axes: xlabel='not_distracted', ylabel='Density'>



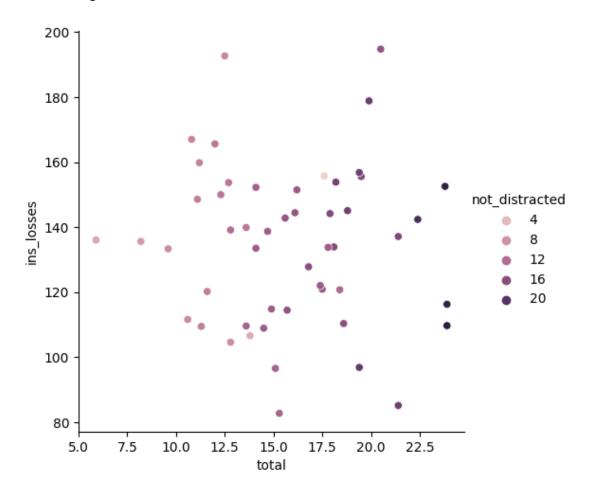
```
[]: #Relationplot sns.relplot(x="total",y="ins_losses",data=ak,hue="abbrev")
```

[]: <seaborn.axisgrid.FacetGrid at 0x7e3d2e3066b0>



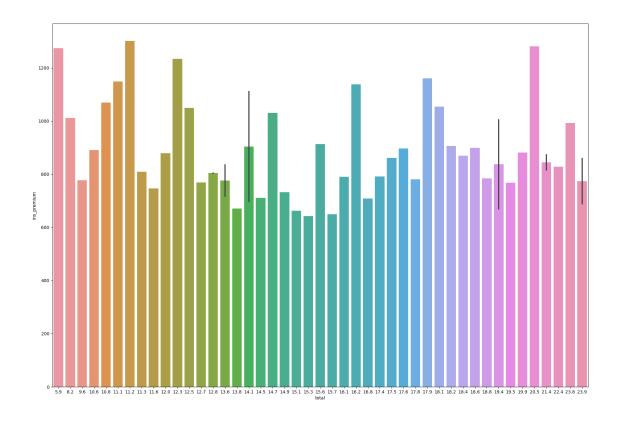
```
[]: #Relationplot
sns.relplot(x="total",y="ins_losses",data=ak,hue="not_distracted")
```

[]: <seaborn.axisgrid.FacetGrid at 0x7e3d2e496fe0>



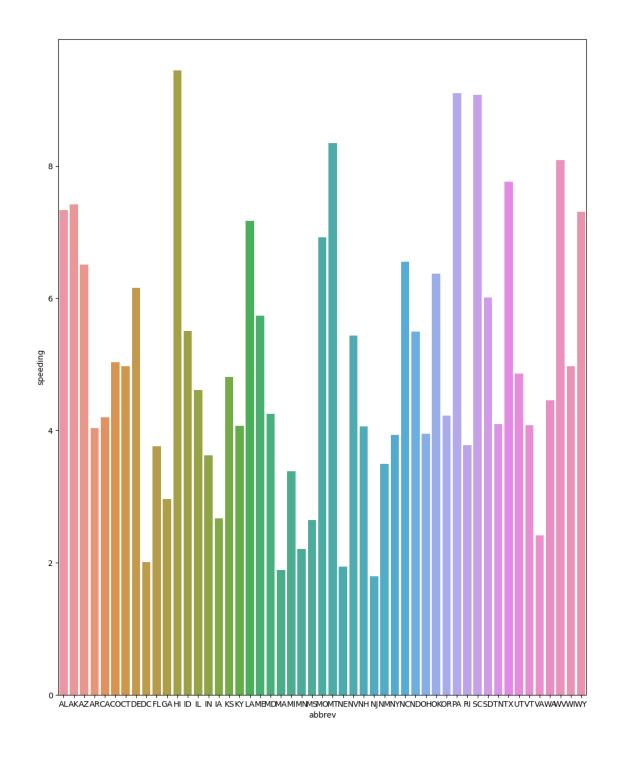
```
[]: #barplot
plt.subplots(figsize=(22,15))
sns.barplot(x="total",y="ins_premium",data=ak)
```

[]: <Axes: xlabel='total', ylabel='ins_premium'>



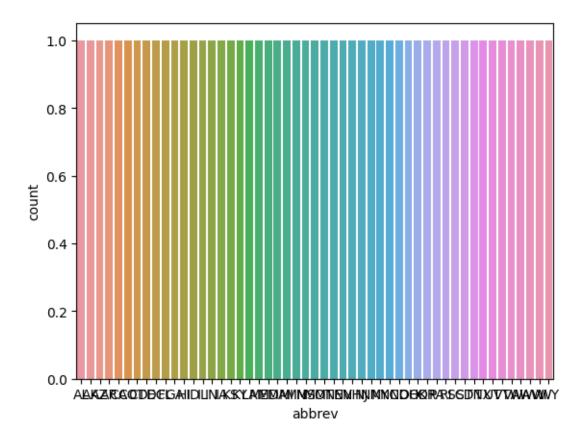
```
[]: #barplot
plt.subplots(figsize=(12,15))
sns.barplot(y="speeding",x="abbrev",data=ak)
```

[]: <Axes: xlabel='abbrev', ylabel='speeding'>



```
[]: #Count Plot
sns.countplot(x="abbrev",data=ak)
```

[]: <Axes: xlabel='abbrev', ylabel='count'>



```
[]: #boxplot
plt.subplots(figsize=(22,15))
sns.boxplot(x="total",y="ins_losses",data=ak)
```

[]: <Axes: xlabel='total', ylabel='ins_losses'>

